```
In [93]: import pandas as pd
In [94]: data=pd.read csv("/home/placement/Downloads/fiat500.csv")
In [95]: data.describe()
Out[951:
                            ID engine power
                                              age in days
                                                                     km previous owners
                                                                                                   lat
                                                                                                               lon
                                                                                                                           price
             count 1538.000000
                                  1538.000000
                                              1538.000000
                                                             1538.000000
                                                                                          1538.000000
                                                                                                       1538.000000
                                                                                                                     1538.000000
                                                                              1538.000000
                     769.500000
                                                                                 1.123537
                                    51.904421
                                               1650.980494
                                                            53396.011704
                                                                                             43.541361
                                                                                                         11.563428
                                                                                                                     8576.003901
             mean
                     444.126671
                                     3.988023
                                               1289.522278
                                                            40046.830723
                                                                                 0.416423
                                                                                              2.133518
                                                                                                          2.328190
                                                                                                                     1939.958641
               std
                      1.000000
                                    51.000000
                                                366.000000
                                                             1232.000000
                                                                                 1.000000
                                                                                             36.855839
                                                                                                          7.245400
                                                                                                                     2500.000000
               min
              25%
                     385.250000
                                    51.000000
                                                670.000000
                                                            20006.250000
                                                                                 1.000000
                                                                                             41.802990
                                                                                                          9.505090
                                                                                                                     7122.500000
              50%
                     769.500000
                                    51.000000
                                               1035.000000
                                                            39031.000000
                                                                                 1.000000
                                                                                             44.394096
                                                                                                         11.869260
                                                                                                                     9000.000000
                                                                                                                   10000.000000
              75%
                    1153.750000
                                    51.000000
                                               2616.000000
                                                            79667.750000
                                                                                 1.000000
                                                                                             45.467960
                                                                                                         12.769040
              max 1538.000000
                                    77.000000
                                               4658.000000
                                                           235000.000000
                                                                                 4.000000
                                                                                             46.795612
                                                                                                         18.365520
                                                                                                                   11100.000000
In [96]:
           data.head()
Out[96]:
                   model engine_power age_in_days
                                                         km previous owners
                                                                                               Ion price
                                                                                     lat
               1 lounge
             0
                                     51
                                                 882
                                                       25000
                                                                            1 44.907242
                                                                                          8.611560
                                                                                                   8900
                2
                      pop
                                     51
                                                1186
                                                       32500
                                                                              45.666359
                                                                                        12.241890
                                                                                                    8800
                     sport
                                     74
                                                4658
                                                     142228
                                                                              45.503300 11.417840
                                                                                                   4200
                                     51
                                                      160000
                                                                              40.633171 17.634609
                   lounge
                                                2739
                                                                                                    6000
                5
                      qoq
                                     73
                                                3074 106880
                                                                            1 41.903221 12.495650
                                                                                                   5700
In [97]: | datal=data.drop(['lat','lon','ID'],axis=1)
```

In [98]: data1

Out[98]:

	model	engine_power	age_in_days	km	previous_owners	price
0	lounge	51	882	25000	1	8900
1	pop	51	1186	32500	1	8800
2	sport	74	4658	142228	1	4200
3	lounge	51	2739	160000	1	6000
4	pop	73	3074	106880	1	5700
1533	sport	51	3712	115280	1	5200
1534	lounge	74	3835	112000	1	4600
1535	pop	51	2223	60457	1	7500
1536	lounge	51	2557	80750	1	5990
1537	pop	51	1766	54276	1	7900

1538 rows × 6 columns

```
In [ ]:
    In [ ]:
    In [99]: data1.shape
    Out[99]: (1538, 6)
In [100]: data1=pd.get_dummies(data1)
```

In [101]: data1

Out[101]

]:		engine_power	age_in_days	km	previous_owners	price	model_lounge	model_pop	model_sport
'	0	51	882	25000	1	8900	1	0	0
	1	51	1186	32500	1	8800	0	1	0
	2	74	4658	142228	1	4200	0	0	1
	3	51	2739	160000	1	6000	1	0	0
	4	73	3074	106880	1	5700	0	1	0
	1533	51	3712	115280	1	5200	0	0	1
	1534	74	3835	112000	1	4600	1	0	0
	1535	51	2223	60457	1	7500	0	1	0
	1536	51	2557	80750	1	5990	1	0	0
	1537	51	1766	54276	1	7900	0	1	0

1538 rows × 8 columns

```
In [102]: data1.shape
Out[102]: (1538, 8)
In [103]: y=data1['price']
x=data1.drop('price',axis=1)
```

```
In [104]: y
Out[104]: 0
                    8900
                    8800
           2
                   4200
           3
                    6000
                    5700
           4
           1533
                   5200
           1534
                   4600
           1535
                   7500
           1536
                    5990
           1537
                   7900
           Name: price, Length: 1538, dtype: int64
In [105]: | from sklearn.model_selection import train_test_split
           x_train, x_test, y_train, y_test =train_test_split(x,y, test_size=0.33,random_state=42)#split data into trai
In [106]: |x_test.head(5)
Out[106]:
                 engine_power age_in_days
                                           km previous_owners model_lounge model_pop model_sport
                                                           2
                         51
                                   3197 120000
                                                                       0
             481
                                                                                1
                                                                                           0
             76
                         62
                                   2101 103000
                                                           1
                                                                       0
                                                                                            0
                                                                                1
            1502
                         51
                                    670
                                         32473
                                                           1
                                                                       1
                                                                                0
                                                                                            0
             669
                         51
                                    913
                                         29000
                                                           1
                                                                       1
                                                                                0
                                                                                            0
            1409
                         51
                                    762
                                         18800
                                                          1
                                                                       1
                                                                                0
                                                                                           0
In [107]: x train.shape
Out[107]: (1030, 7)
```

```
In [108]: y train
Out[108]: 527
                    9990
           129
                    9500
           602
                    7590
           331
                    8750
           323
                    9100
           1130
                   10990
          1294
                    9800
          860
                     5500
          1459
                    9990
          1126
                    8900
          Name: price, Length: 1030, dtype: int64
In [109]: y_test.head(5)
Out[109]: 481
                   7900
                   7900
           76
           1502
                   9400
           669
                   8500
           1409
                   9700
          Name: price, dtype: int64
In [110]: y_train.shape
Out[110]: (1030,)
In [111]: from sklearn.linear model import LinearRegression
           reg=LinearRegression() #creating object of LinearRegression
          reg.fit(x_train,y_train)#training and fitting LR object using training data
Out[111]: LinearRegression()
          In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
          On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
In [112]: ypred=reg.predict(x test)
```

```
In [113]: ypred
Out[113]: array([ 5867.6503378 ,
                                  7133.70142341.
                                                   9866.35776216.
                                                                   9723.28874535.
                 10039.59101162,
                                  9654.07582608,
                                                   9673.14563045, 10118.70728123,
                  9903.85952664,
                                  9351.55828437, 10434.34963575, 7732.26255693,
                                  6565.95240435,
                                                   9662.90103518, 10373.20344286,
                  7698.67240131,
                  9599.94844451,
                                  7699.34400418,
                                                   4941.33017994, 10455.2719478 ,
                 10370.51555682, 10391.60424404,
                                                   7529.06622456,
                                                                   9952.37340054,
                  7006.13845729,
                                  9000.1780961 ,
                                                   4798.36770637,
                                                                   6953.10376491,
                                                   7333.52158317,
                  7810.39767825,
                                  9623.80497535,
                                                                   5229.18705519,
                  5398.21541073,
                                  5157.65652129,
                                                   8948.63632836,
                                                                   5666.62365159,
                                                                   8457.38443276,
                  9822.1231461 ,
                                  8258.46551788,
                                                   6279.2040404 ,
                  9773.86444066,
                                  6767.04074749,
                                                   9182.99904787, 10210.05195479,
                  8694.90545226, 10328.43369248,
                                                                   8866.7826029 ,
                                                   9069.05761443,
                  7058.39787506,
                                  9073.33877162,
                                                   9412.68162121, 10293.69451263,
                 10072.49011135,
                                  6748.5794244 ,
                                                   9785.95841801,
                                                                   9354.09969973,
                  9507.9444386 , 10443.01608254,
                                                   9795.31884316,
                                                                   7197.84932877,
                 10108.31707235, 7009.6597206,
                                                   9853.90699412,
                                                                   7146.87414965,
                                                                   8515.83255277,
                  6417.69133992,
                                  9996.97382441,
                                                   9781.18795953,
                  8456.30006203,
                                                                   6832.86406122,
                                  6499.76668237,
                                                   7768.57829985,
                                                                   8562.56562053.
                  8347.96113362, 10439.02404036,
                                                   7356.43463051,
In [114]: from sklearn.metrics import r2 score
          r2 score(y test,ypred)
Out[114]: 0.8415526986865394
In [116]: from sklearn.metrics import mean squared error #calculating MSE
          mean squared error(ypred,y test)
Out[116]: 581887.727391353
In [117]:
          n=581887.727391353
          print(n**(1/2))
          762.8156575420782
```

```
In [118]: y test.head(5)
Out[118]: 481
                   7900
          76
                   7900
          1502
                   9400
          669
                   8500
          1409
                   9700
          Name: price, dtype: int64
In [121]:
          vpred
                                   7133.70142341,
Out[121]: array([ 5867.6503378 ,
                                                     9866.35776216,
                                                                     9723.28874535,
                                    9654.07582608,
                                                     9673.14563045, 10118.70728123,
                  10039.59101162,
                   9903.85952664,
                                   9351.55828437, 10434.34963575,
                                                                     7732.26255693,
                   7698.67240131,
                                    6565.95240435,
                                                     9662.90103518, 10373.20344286,
                   9599.94844451,
                                                    4941.33017994, 10455.2719478 ,
                                   7699.34400418,
                  10370.51555682, 10391.60424404,
                                                    7529.06622456,
                                                                     9952.37340054,
                   7006.13845729,
                                    9000.1780961 ,
                                                     4798.36770637,
                                                                     6953.10376491,
                   7810.39767825,
                                    9623.80497535,
                                                    7333.52158317,
                                                                     5229.18705519,
                   5398.21541073,
                                                                     5666.62365159,
                                   5157.65652129,
                                                     8948.63632836,
                   9822.1231461 ,
                                    8258.46551788,
                                                     6279.2040404 ,
                                                                     8457.38443276,
                   9773.86444066,
                                   6767.04074749,
                                                     9182.99904787, 10210.05195479,
                   8694.90545226, 10328.43369248,
                                                     9069.05761443,
                                                                     8866.7826029 ,
                   7058.39787506,
                                   9073.33877162,
                                                     9412.68162121, 10293.69451263,
                  10072.49011135,
                                   6748.5794244 ,
                                                     9785.95841801,
                                                                     9354.09969973,
                   9507.9444386 ,
                                  10443.01608254,
                                                     9795.31884316,
                                                                     7197.84932877,
                  10108.31707235,
                                   7009.6597206 ,
                                                     9853.90699412,
                                                                     7146.87414965,
                                    9996.97382441,
                                                                     8515.83255277,
                   6417.69133992,
                                                     9781.18795953,
                   8456.30006203,
                                   6499.76668237,
                                                     7768.57829985,
                                                                     6832.86406122,
                   8347.96113362, 10439.02404036,
                                                     7356.43463051,
                                                                     8562.56562053,
                   0020 70555100
                                   10025 02571520
                                                                     0411 45004006
                                                     7270 77100022
```

```
In [124]: Results=pd.DataFrame(columns=['Price', 'Predicted'])
    Results['Price']=y_test
    Results['Predicted']=ypred
    Results.head(15)
```

Out[124]:		Price	Predicted
	481	7900	5867.650338
	76	7900	7133.701423
	1502	9400	9866.357762
	669	8500	9723.288745
	1409	9700	10039.591012
	1414	9900	9654.075826
	1089	9900	9673.145630
	1507	9950	10118.707281
	970	10700	9903.859527
	1198	8999	9351.558284
	1088	9890	10434.349636
	576	7990	7732.262557
	965	7380	7698.672401
	1488	6800	6565.952404

```
In [129]: Results['diff']=Results.apply(lambda row: row.Price - row.Predicted,axis=1)
```

1432

8900

9662.901035

In [130]: Results

Out[130]:

	Price	Predicted	diff
481	7900	5867.650338	2032.349662
76	7900	7133.701423	766.298577
1502	9400	9866.357762	-466.357762
669	8500	9723.288745	-1223.288745
1409	9700	10039.591012	-339.591012
291	10900	10032.665135	867.334865
596	5699	6281.536277	-582.536277
1489	9500	9986.327508	-486.327508
1436	6990	8381.517020	-1391.517020
575	10900	10371.142553	528.857447

508 rows × 3 columns

In [ ]: